



NASA'S TECHNOLOGY INFUSION  
**ROAD TOUR**

*Historically Black Colleges/Universities & Minority Serving Institutions*

**NEW MEXICO STATE UNIVERSITY**

**New Mexico Highlands University  
AI/Machine Learning in Materials  
Science**

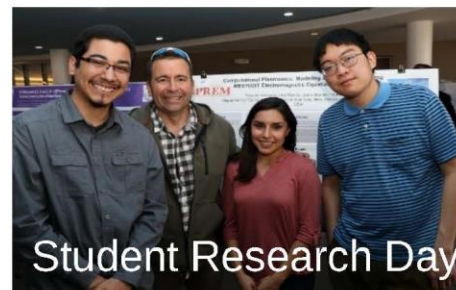
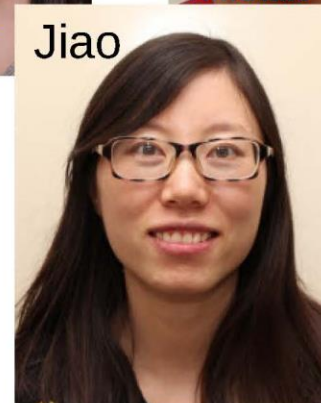
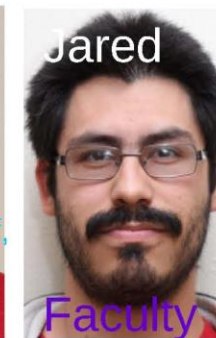
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**505-414-1562**

**Date 8-13-19**



# Research - Thrust 3b: Machine learning - artificial intelligence - material sciences





**Objective:** Characterization of biosensor morphology utilizing machine learning to predict resonant frequency of novel scanning electron microscope (SEM) images of random nanostar distributions.

**Step 1:** Novel SEM image is pre-processed by Feature Identification Algorithm (FIA)

**Step 2:** Construct feature vector with features  $f_i$ ,  $i=1,2,3,...m$

$$X = \begin{bmatrix} x1 \\ x2 \\ \dots \\ xn \end{bmatrix} = [\text{Features}]$$

$$\begin{aligned} x1 &= [f1 \ f2 \ f3...fm] \\ x2 &= [f1 \ f2 \ f3...fm] \\ \dots & \\ xn &= [f1 \ f2 \ f3...fm] \end{aligned}$$

**Step 3:** Utilizing FIA information –construct electromagnetic models for MIT Electromagnetic Equation Propagation (MEEP) solver.

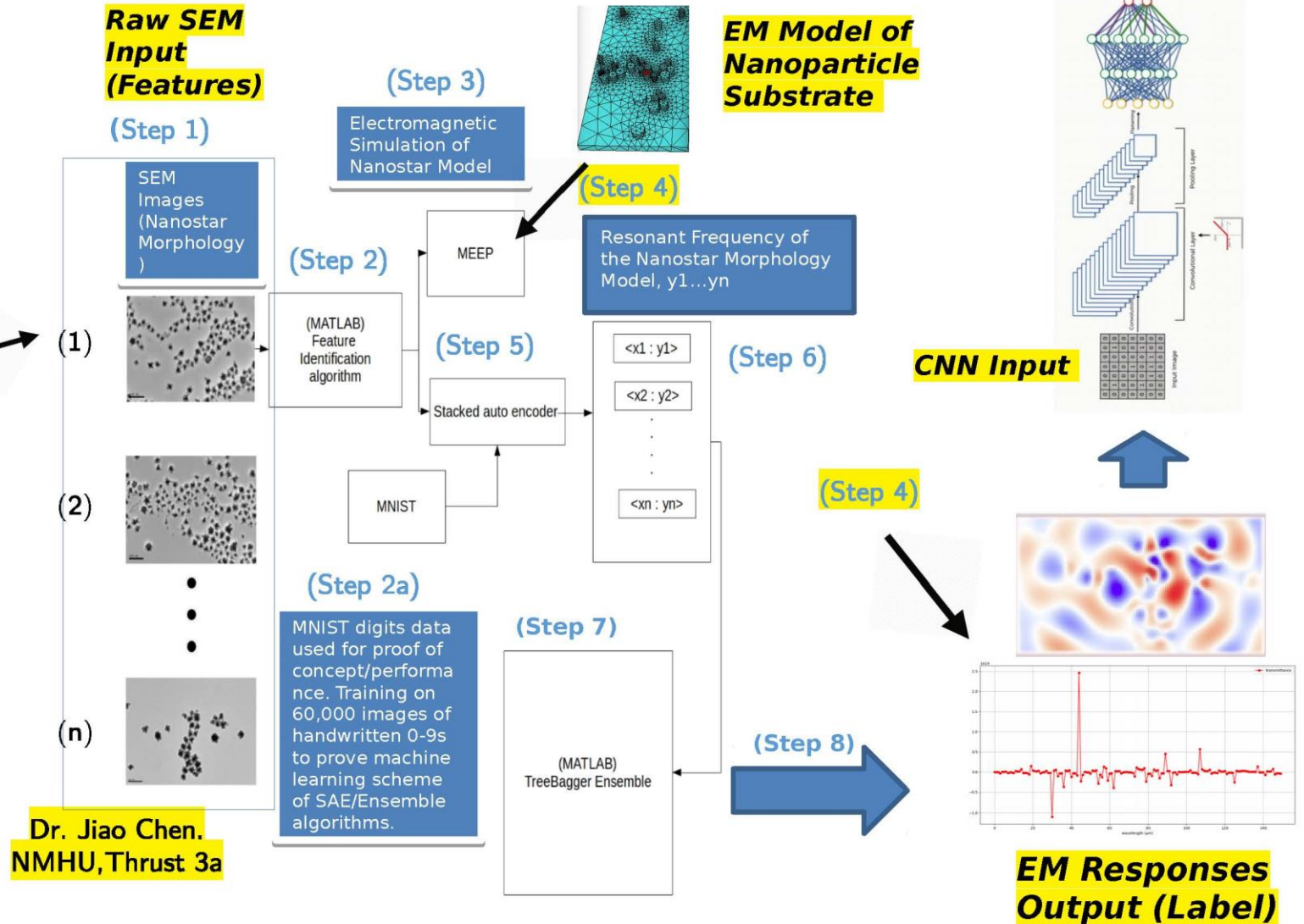
**Step 4:** Solver computes resonant frequency for SEM morphology,  $y_i$ ,  $i=1,2,3...n$

**Step 5:** Stacked Auto-Encoder (SAE) learns lower dimensional vector  $h_i$  to represent  $x_i$  for  $i=1,2,3,...n$

**Step 6:** New X is created from  $h_i$  and  $y_i$

**Step 7:** Ensemble, machine learning algorithm uses  $h_i$  features and  $y_i$ , predictions to build prediction model for resonant frequency of SEM morphologies.

**Step 8:** Novel SEM image (not in training set) is then processed Steps 1 through 7. Resonant frequency is predicted (Excitation light wavelength now known)







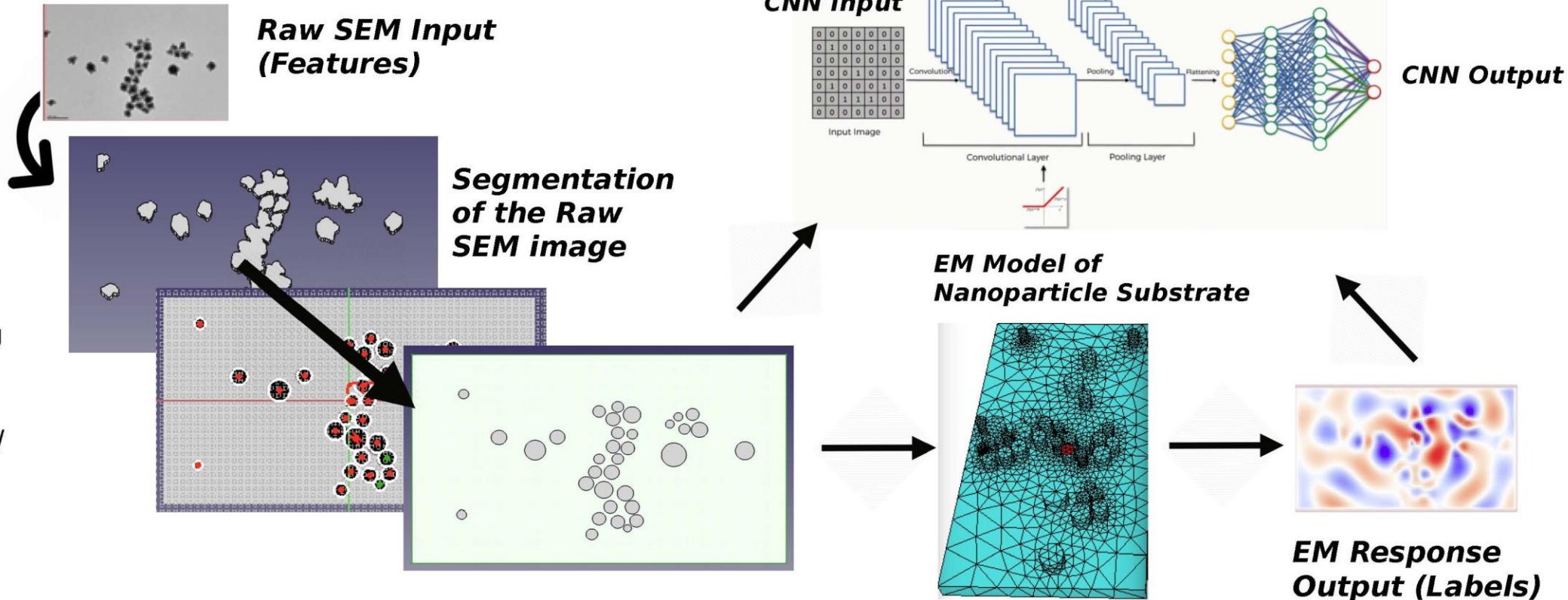
Henry Chung  
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2<sup>nd</sup> Semester Graduate  
Student , MS CS, NMHU

**Research:**  
Computational modeling/  
simulation of  
metamaterials and  
AI/machine learning

**Employment:** PREM

# Research - Project 2: **Generation of Machine Learning Output (Label) Images for Convolutional Neural Network (CNN) Training :Utilizing Ab Initio Electromagnetic (EM) Simulation of Nanoparticle Substrate** Input (Features) from Scanning Electron Microscope (SEM) Images







**Fred Sena**  
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**Research:** Computational  
modeling/ simulation of  
metamaterials and  
AI/machine learning

**Employment:** LANL

# Research - Project 3: *Predicting Atomization Energies Using Machine Learning And Principal Component Analysis (PCA)*



- Machine Learning and Principal Component Analysis (PCA)
- Predict atomization energies of crystal structures
- Dataset containing over 7000 molecules
- Dataset contained over 500 features
- PCA is used to extract only the most important features
- Reduces the original feature set by nearly 70%
- The primary goal of this research is to improve the computational efficiency of the machine learning model, while ensuring the accuracy is not lost

```
model = Sequential()
model.add(Dense(256, activation=keras.activations.linear)),
model.add(Dense(128, activation=keras.activations.linear)),
model.add(Dense(64, activation=keras.activations.linear)),
model.add(Dropout(0.1)),
model.add(Dense(1, activation=keras.activations.linear))

model.compile(optimizer='adam',
              loss='mean_squared_error',
              metrics=['accuracy'])

model.fit(x_train, y_train, epochs=200)
model.evaluate(x_test, y_test)
y_pred = model.predict(x_test)
```

**Python, KERAS ,  
Neural Network  
Code Snippet**

Network	Accuracy	Training Time	Dataset Shape
Neural Network (No PCA)	97.96%	67 Seconds	7165, 529
Neural Network (PCA)	97.87%	49 Seconds	7165, 160

**Results**



**Rosario Vasquez**  
BS CS, NMHU

4<sup>th</sup> Semester Graduate  
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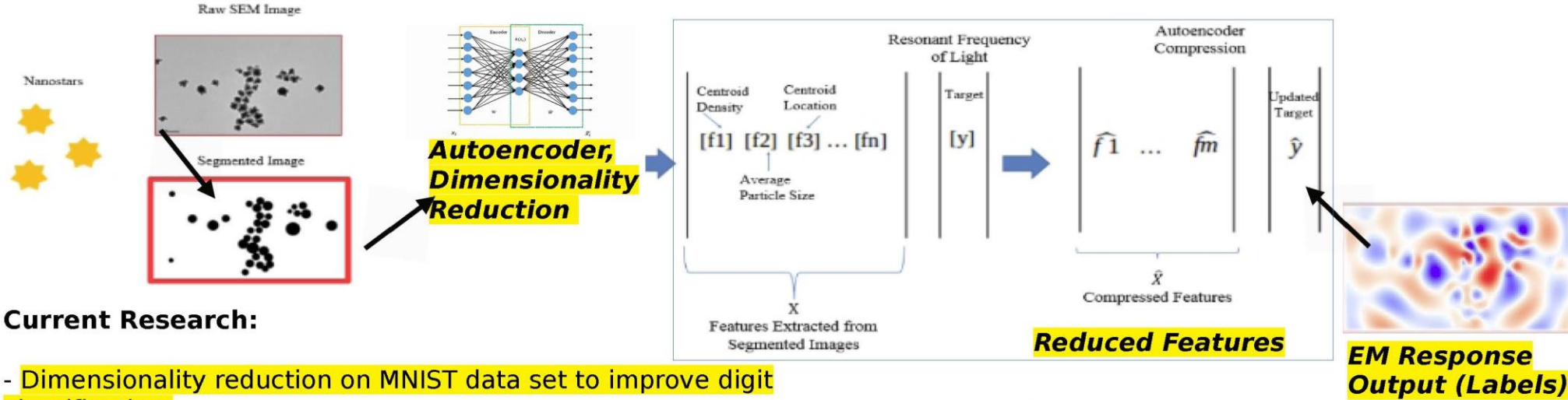
**Research:**  
Computational modeling/  
simulation of  
metamaterials and  
AI/machine learning

**Employment:** LANL

# Research - Project 4:

## Classification/Dimensionality Reduction (DR)

### Utilizing Autoencoders and Ensemble Learning

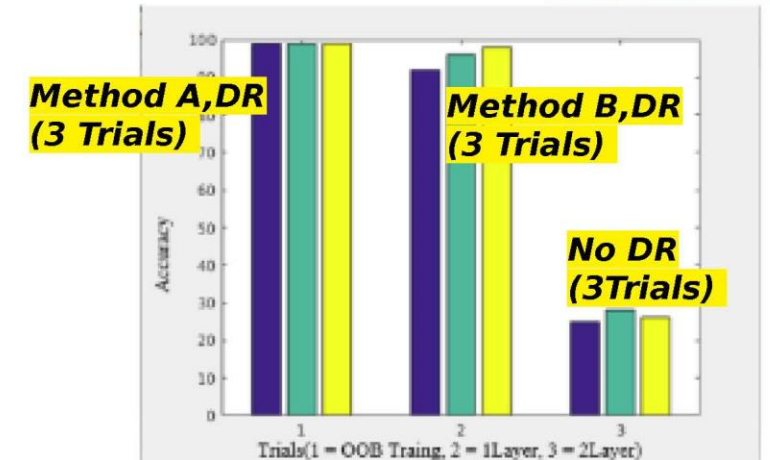


#### Current Research:

- Dimensionality reduction on MNIST data set to improve digit classification
- Autoencoder(AE) learns a reduced manifold model of the input features
- Classifies with a reduced set of features and reduced computational time

#### Future Work:

- Coupling ML methods with finite difference time domain (FDTD) plasmonic simulation methods (See Projects 1 and 2)
- Characterization and prediction of the response of the heterogeneous nanostar substrate morphology to photonic excitation without computationally expensive FDTD simulation of substrate







**Luis Garcia**  
BS CS, NMHU

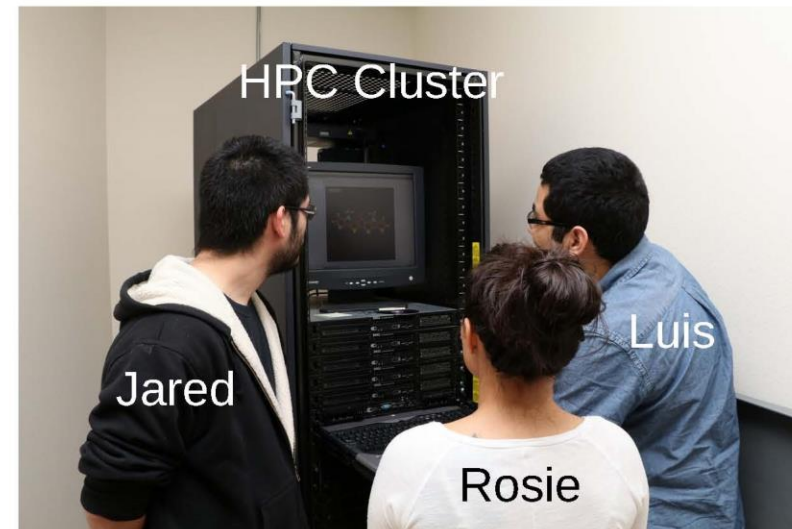
4th Semester Graduate  
Student , MS CS, NMHU

**Research:** Computational  
modeling/ simulation of  
metamaterials and  
AI/machine learning and  
HPC

**Employment:** LANL

# Research - Project 5: Predicting Quantum Mechanical Properties of Molecular Structures Utilizing Density Functional Theory and High Performance Computing

- Predict quantum mechanical properties of molecular structure
- Simulation complex molecules using DFT calculations on a HPC system
- Two parallel computing HPC clusters running Gauss and CUDA
- The primary goal of this research is to improve the computational efficiency of the DFT calculations, while ensuring the accuracy is not lost
- Future work will be to construct features and labels for machine learning and predict quantum mechanical properties using an ANN



Cluster	GPU Cores	CPU Cores	OS
Xeon, 2- Nvidia Tesla K80	~10000	3	Redhat
Xeon, 2-Nvidia Tesla K10	~7000	3	Redhat

## Research - Other Projects

- Development of a novel machine learning methods to take SEM images and EM simulations to generate features and labels for an ANN to train on for prediction of novel SEM image EM response.
- Co-authored and published one refereed paper in IEEE journal related to machine learning and plasmonic biosensors for detection of early onset Alzheimer's Disease.

## Education

- Course Development:
- Data Science I, Data Science II, Machine Learning, Artificial Neural Networks, Artificial Intelligence, Cognitive Science have been offered since 2015 as part of the support for AI and Machine Learning in computer science as it relates to many science disciplines. These courses all included materials science related problems to emphasize the importance in this field.
- A Machine Learning Short Course was taught by experts in industry to twenty undergraduate and graduate students during the summer of 2016. Each student was able to develop skills in solving problems in materials science related to biosensor design utilizing the latest method in machine learning.

## Role of the partnering institution

- Three undergraduate students, Jason Marquez, Patricio Herrera and Taylor Herrera worked in three groups at OSU during the summers from 2017-18. They worked on modeling magnetic materials systems with several of our colleagues at OSU.